

Physical Model based SWE Retrieval Algorithm Using X- and Ku- band Radar Backscatter

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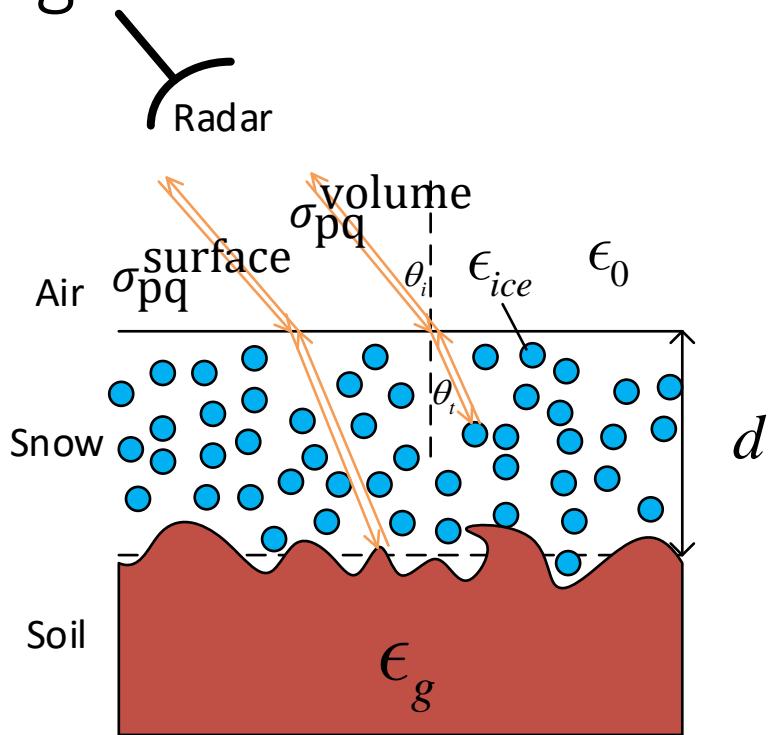
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Outlines

- A. Background scattering subtraction
- B. Forward model:
 - i. Bicontinuous / DMRT model and regression training
 - ii. Parameterized model: only 2 parameters ω_X and τ_X
- C. Physical model based SWE retrieval algorithm
 - i. Radar retrieval algorithm
 - ii. Classify backscatter w.r.t. ω_X
 - iii. SWE retrieval performance Using SnowSAR backscatter σ_{VV} (9.6 GHz and 17.2GHz)

Radar backscattering: volume and surface scattering

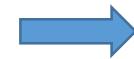


Poster 20

Shurun Tan et al., "Assessment of Background Scattering at X- and Ku-band in Snow Remote Sensing".

$$\sigma_{pq}^{\text{total}} = \sigma_{pq}^{\text{volume}} + \sigma_{pq}^{\text{surface}} \exp\left(-\frac{2\tau}{\cos \theta_t}\right)$$

$\sigma_{pq}^{\text{volume}}$: volume scattering from snowpack

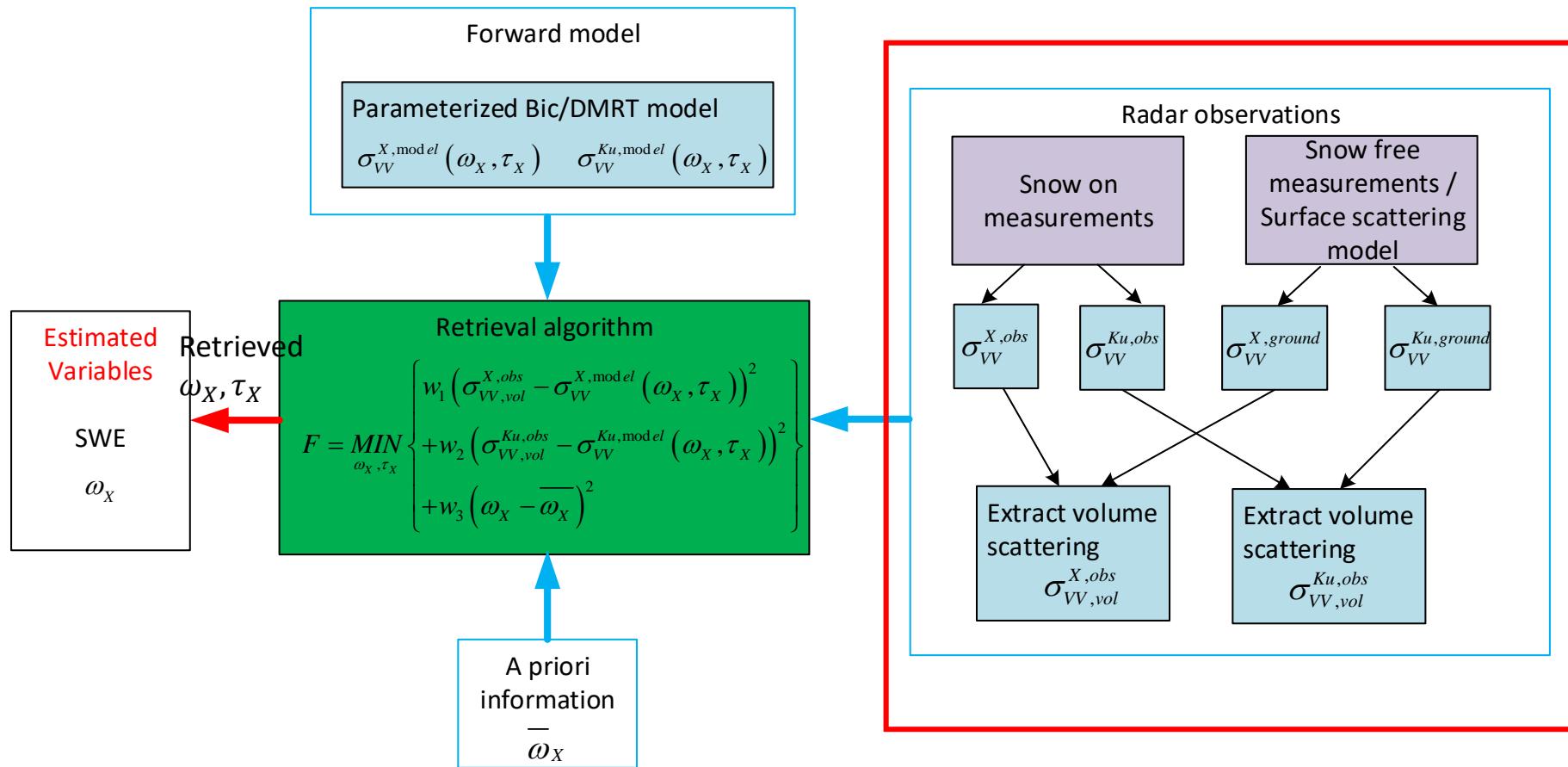


Give SWE

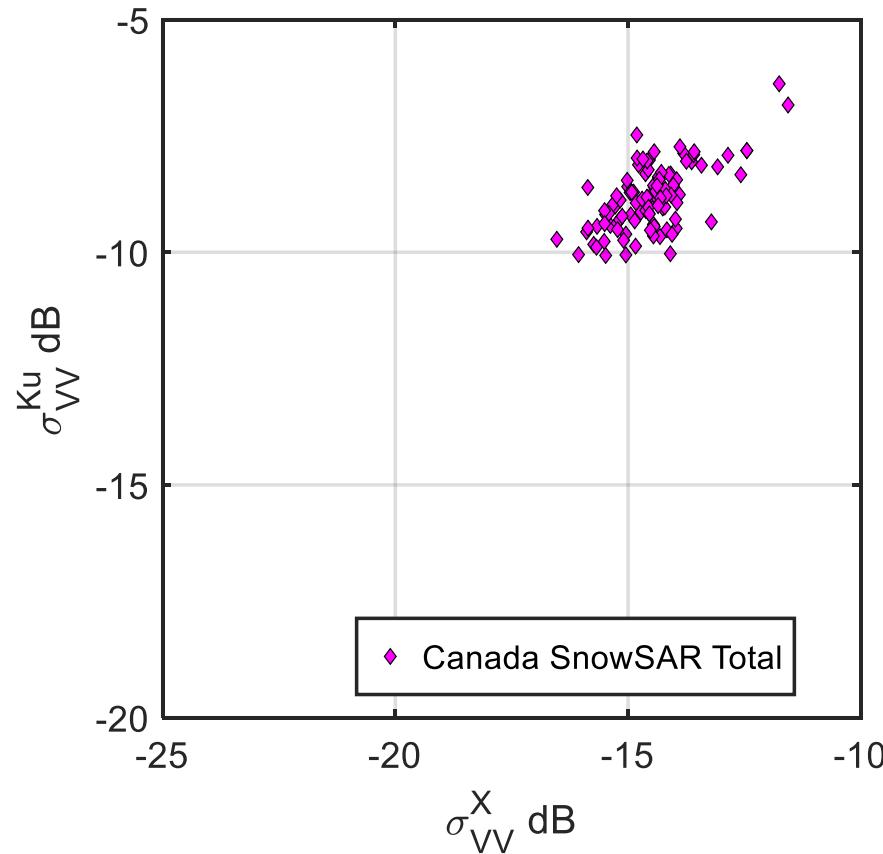
$\sigma_{pq}^{\text{surface}}$: surface scattering from ground



Background scattering subtraction in the SWE retrieval algorithm

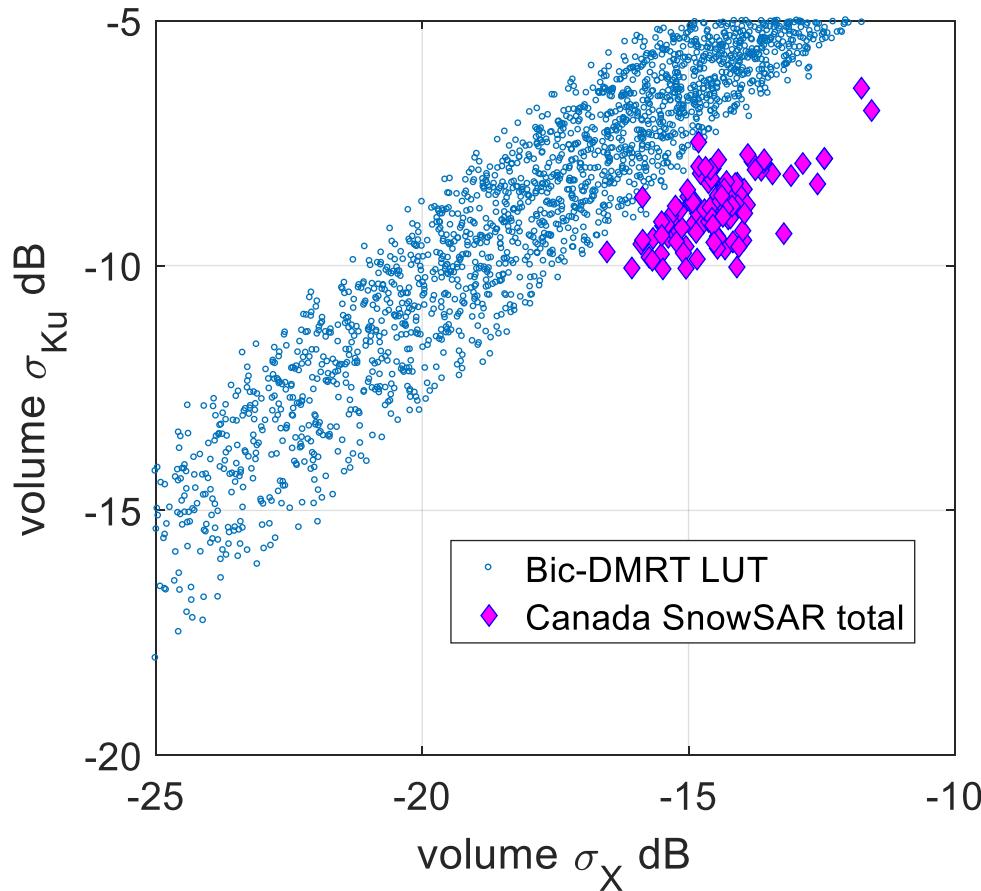


SnowSAR (Canada TVC 2013) X- and Ku-band backscatter: raw data



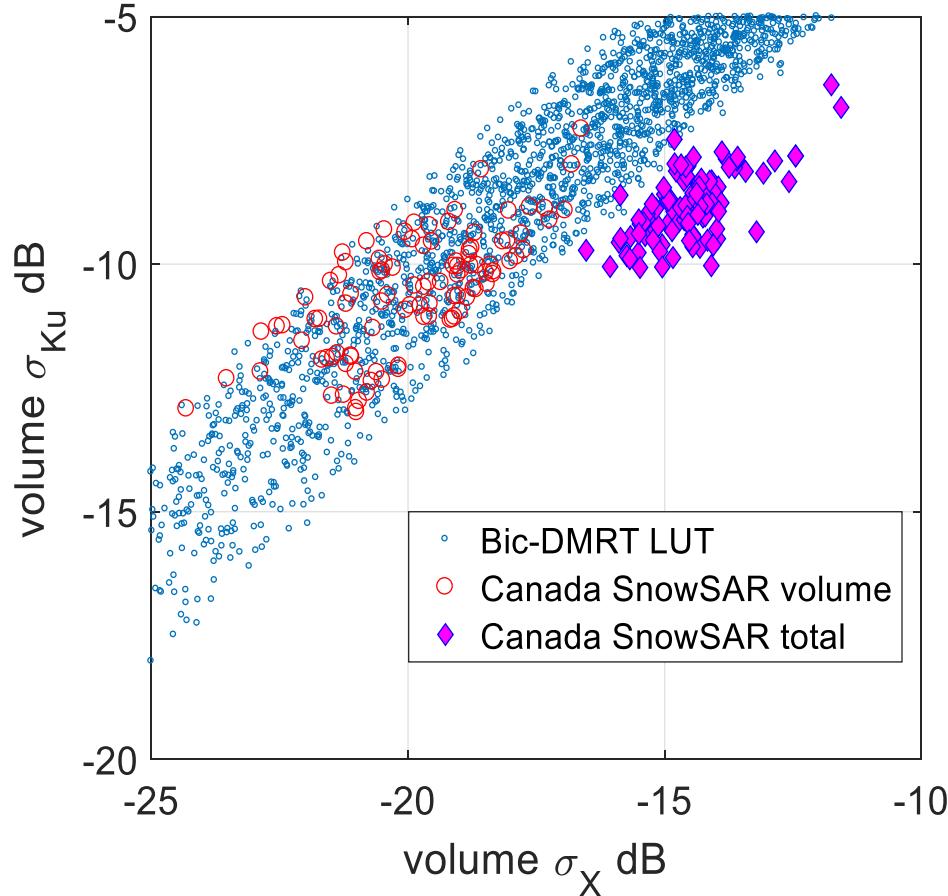
- σ_{VV}^X : ranged from -18dB to -11dB
- σ_{VV}^{Ku} : ranged from -11dB to -6dB

Bic/DMRT LUT compare with Canada SnowSAR



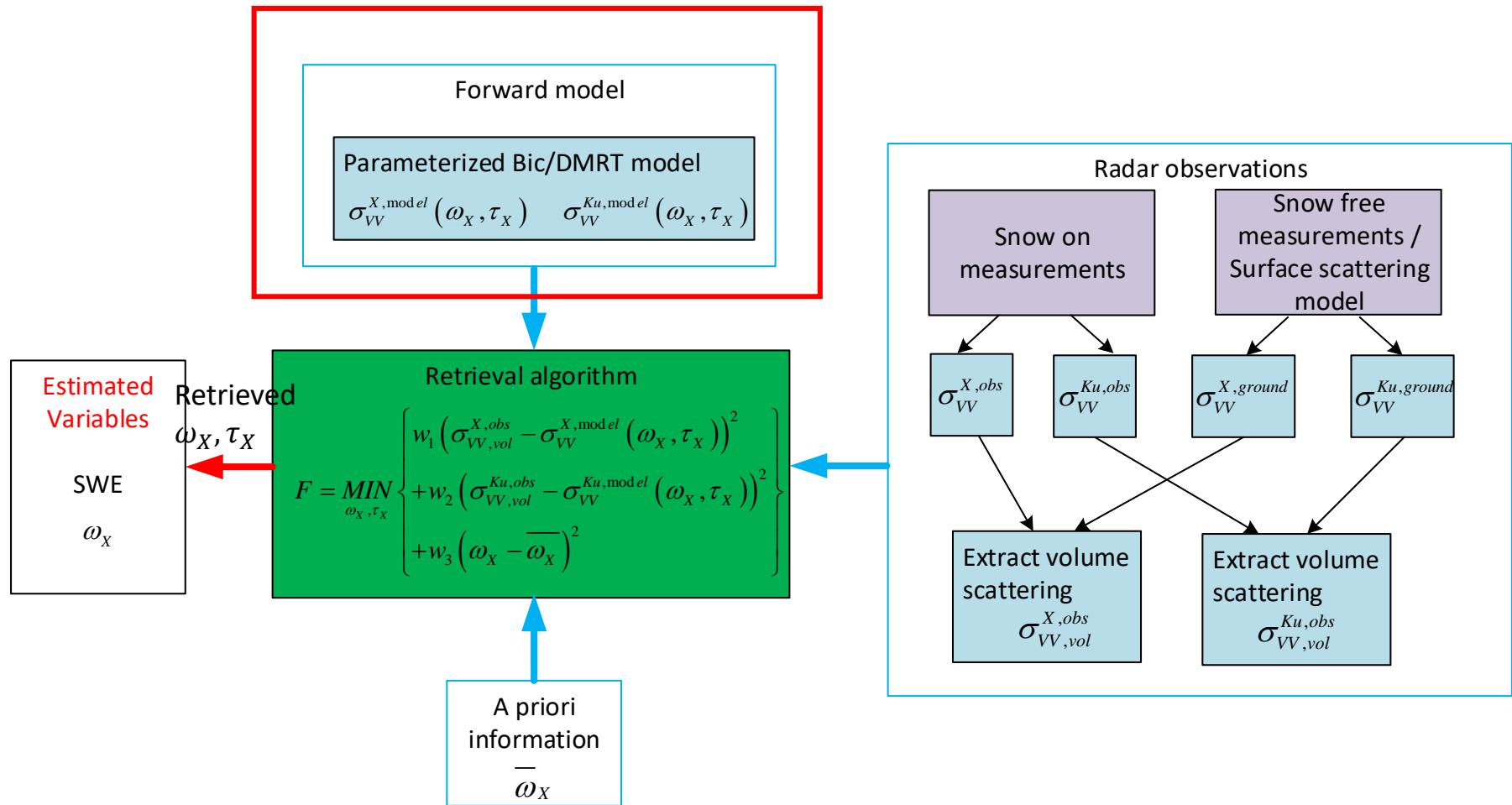
- Model: volume scattering
- SnowSAR data: volume scattering + background scattering

Background scattering subtracted from raw data

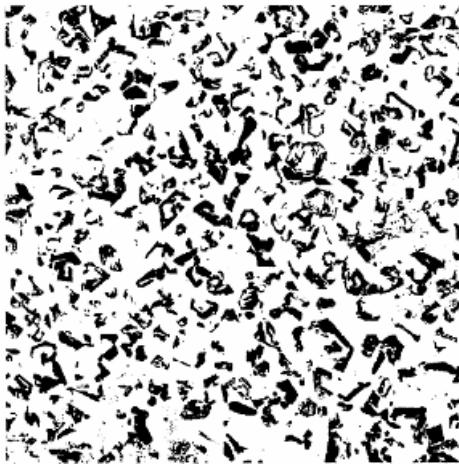


- Volume scattering of SnowSAR within model predictions
- Shift data more in X band than Ku band
- Larger dynamic range in volume scattering

SWE retrieval algorithm flow chart

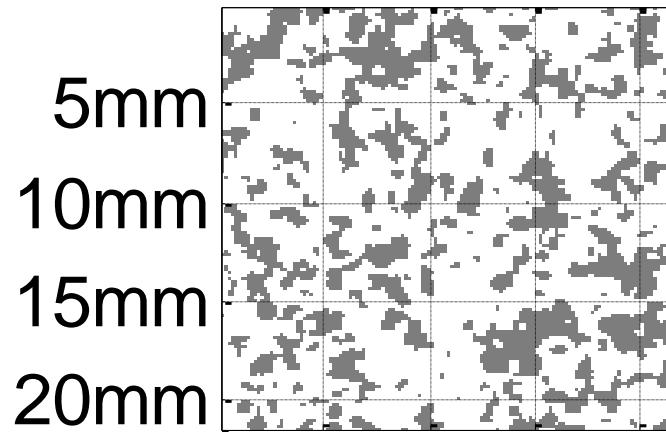


Computer Generated Snow: Bicontinuous Medium



A. Wiesmann, C. Mätzler, and T. Weise, "Radiometric and structural measurements of snow samples," *Radio Sci.*, vol. 33, pp. 273-289, 1998.

Real snow cross
section image

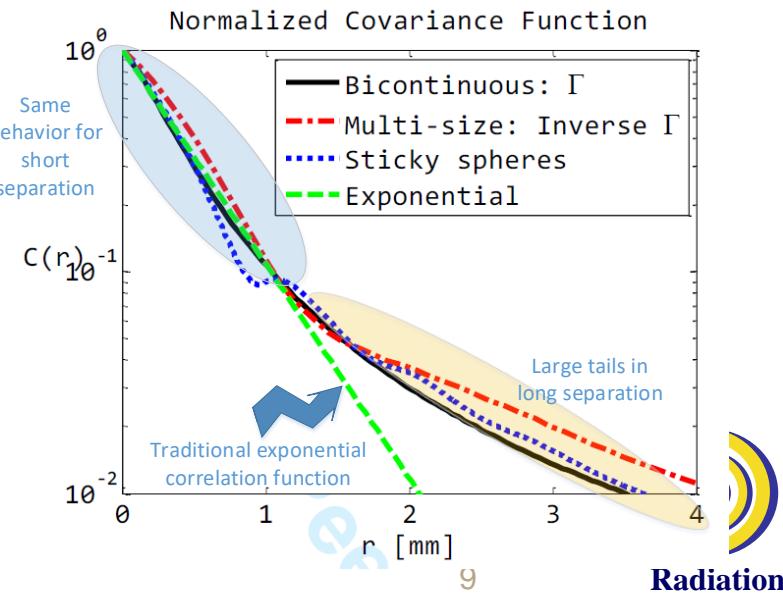


Computer-generated

Comparison through
correlation function

Poster 7

Weihui Gu et al., "DMRT
Models for Active and
Passive Microwave
Remote Sensing"



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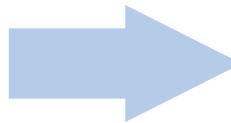


Radiation
Laboratory

Snow homogeneous: Bicontinuous Dense Media Radiative Transfer (Bic/DMRT)

coherent

Solve Maxwell's Eq. over a block of computer snow ($3\lambda - 5\lambda$) with DDA:
get effective $P, \kappa_e, \epsilon_{\text{eff}}$

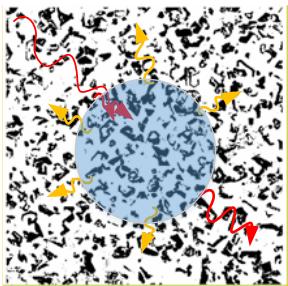


incoherent

Substitute the effective parameters into & Solve RTE:
Backscatter: σ

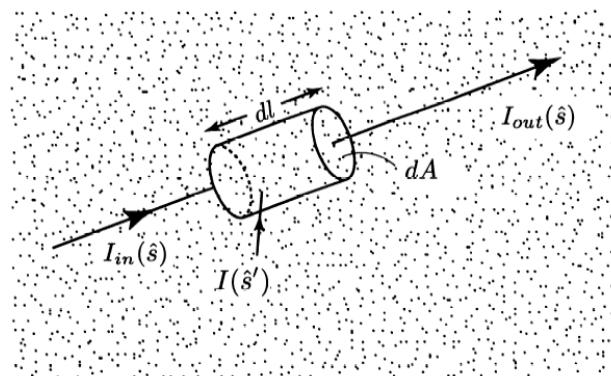
Discrete Dipole Approximation (DDA)

$$\bar{E}(\bar{r}_i) = \bar{E}_{inc}(\bar{r}_i) + \frac{k^2}{\epsilon} \sum_{j=1}^N \bar{\bar{G}}(\bar{r}_i, \bar{r}_j) \cdot \Delta V_j (\epsilon_r(\bar{r}_j) - 1) \bar{E}(\bar{r}_j)$$



Radiative Transfer Equation

$$\frac{dI(\hat{s})}{ds} = -\kappa_e I(\hat{s}) + \int d\hat{s}' P(\hat{s}, \hat{s}') I(\hat{s}')$$



$P(\hat{s}, \hat{s}')$: phase matrix

κ_e : extinction coefficient

$I(\hat{s})$: Intensity in direction \hat{s}

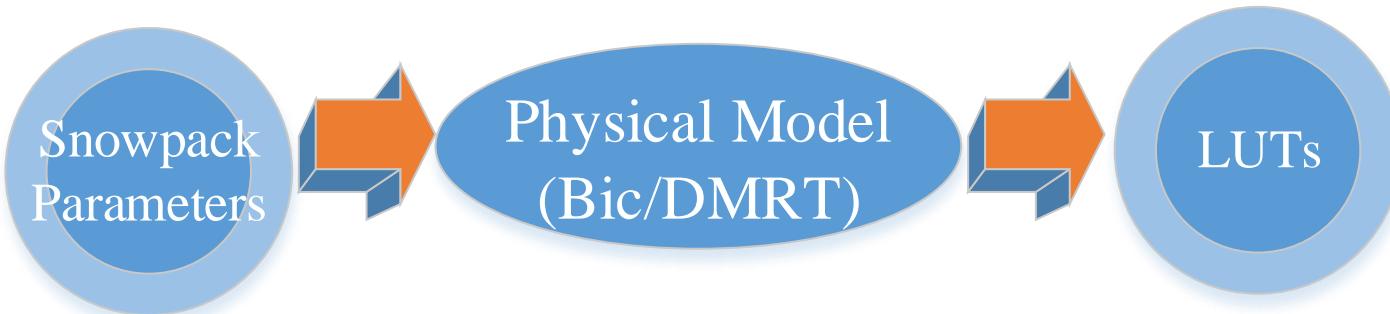


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Poster 7
Weihui Gu et al.


Radiation
Laboratory

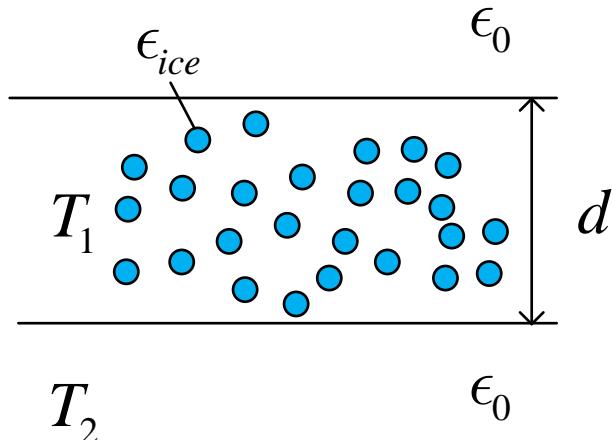
Look-up table (LUT) of Bic/DMRT



Parameters	Minimum	Maximum	Interval
Volume fraction f_v	10%	45%	5%
b parameter	0.6	1.6	0.2
$\langle \zeta \rangle$ parameter (m^{-1})	5000	15000	2000
Snow depth d (m)	0.1	1.2	0.1

SWE	$(\langle \zeta \rangle, b, \rho_{snow}, d)$	$(\sigma_{VV}^X, \sigma_{VV}^{Ku})$ dB	ω_X	τ_X	...
55.02	(9000, 1.2, 10%, 0.6)	(-15.3, -10.6)	0.6805	0.0166	...
64.19	(9000, 1.2, 10%, 0.7)	(-14.9, -10.1)	0.6805	0.0194	...
73.36	(9000, 1.2, 10%, 0.8)	(-14.6, -9.7)	0.6805	0.0221	...
...

Parameterization: scattering albedo ω and optical thickness τ , retrieve τ_a



κ_s : scattering coefficients
 κ_a : absorption coefficients
 κ_e : extinction coefficients

$$\kappa_e = \kappa_a + \kappa_s$$

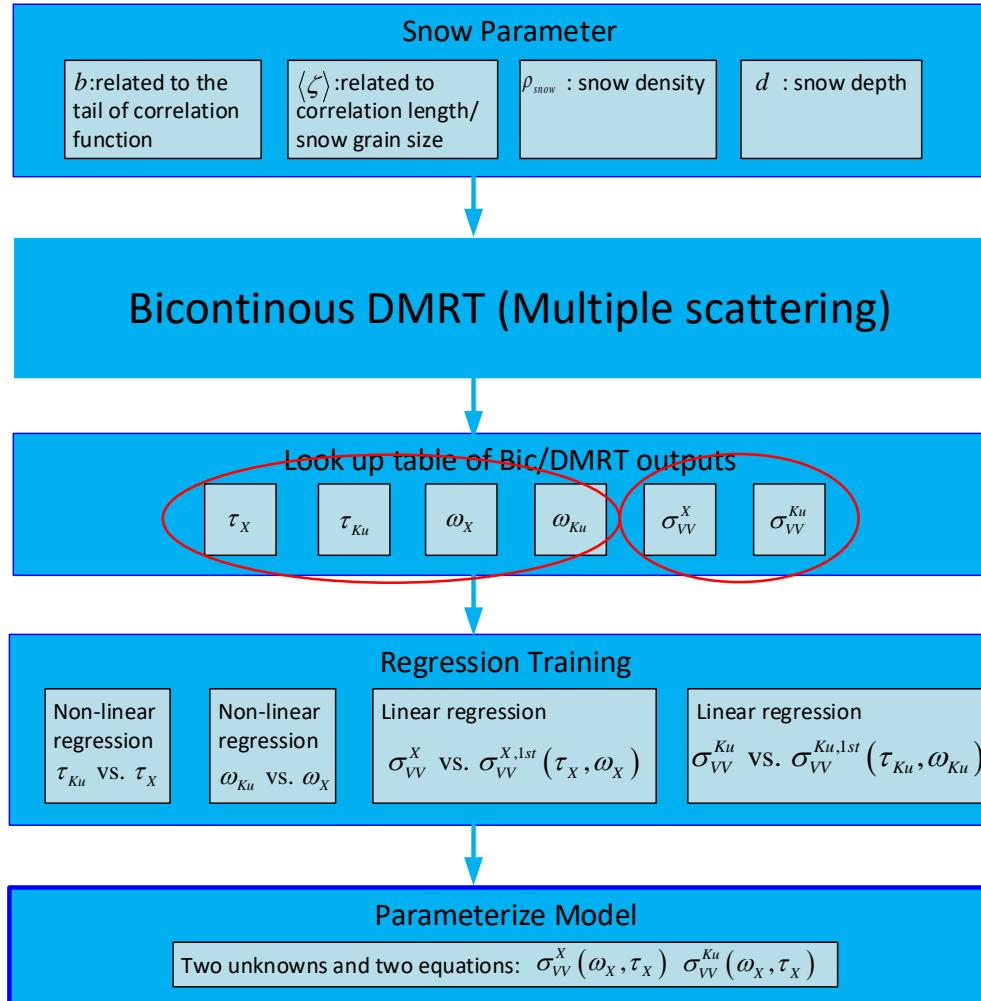
- ❑ Scattering albedo: $\omega = \frac{\kappa_s}{\kappa_s + \kappa_a} = \frac{\kappa_s}{\kappa_e}$
- ❑ Optical thickness: $\tau = \kappa_e d$

- ❑ Absorption loss is proportional to SWE

$$\tau_a = (1 - \omega)\tau = \kappa_a d \propto \text{SWE}$$

- ❑ Two frequencies, four parameters: $\omega_X, \tau_X; \omega_{\text{Ku}}, \tau_{\text{Ku}}$

Regression training: reduce $\omega_{Ku}, \omega_X, \tau_{Ku}, \tau_X$ to ω_X, τ_X

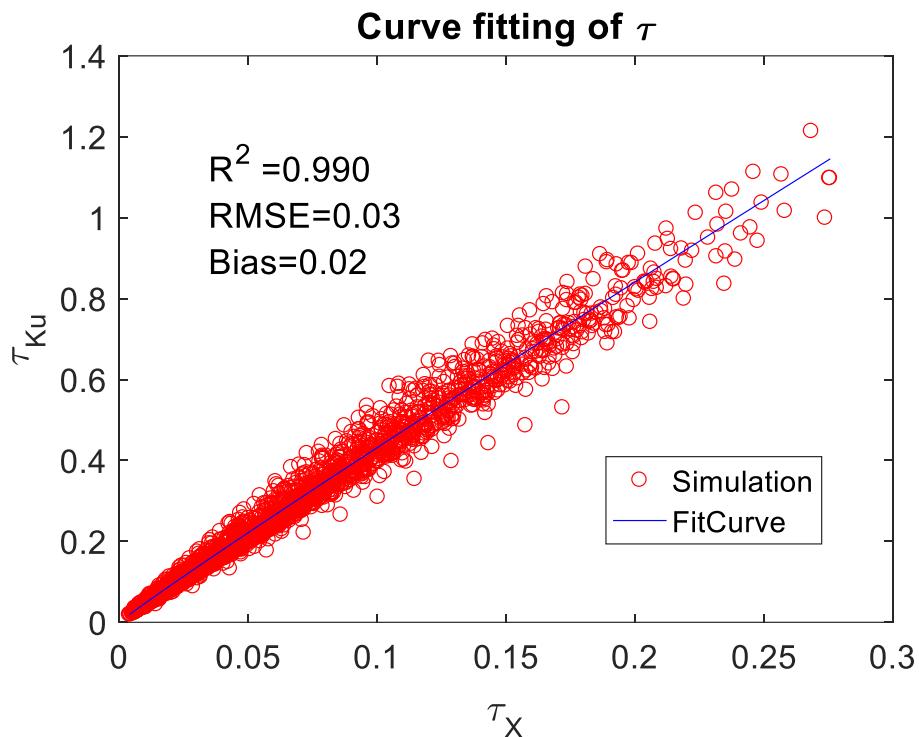


- ❑ Four parameters
- ❑ Two observations

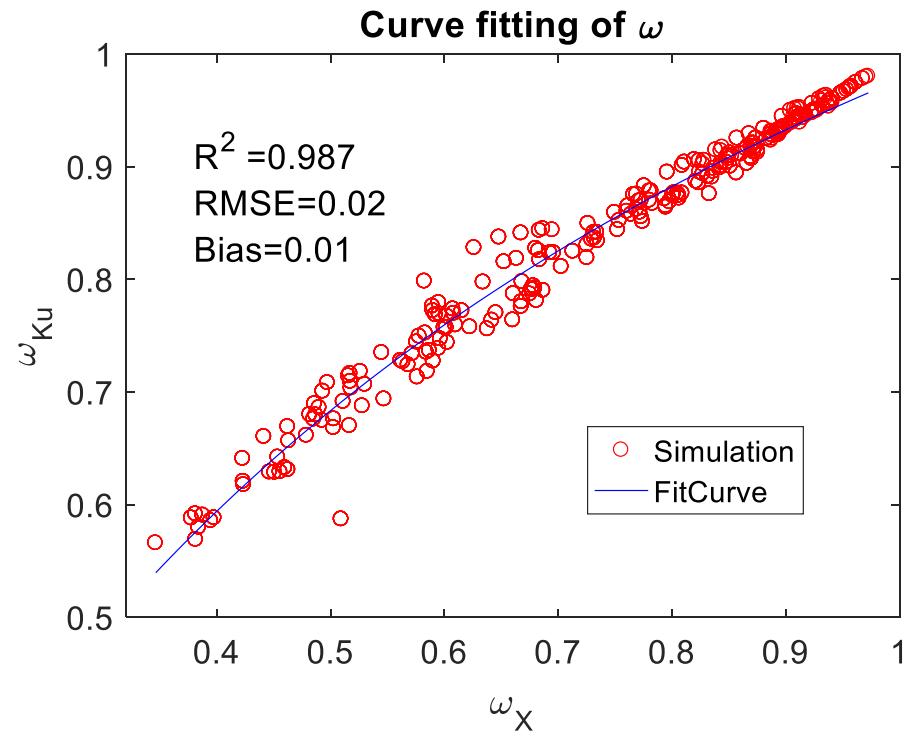
- ❑ Two parameters
- ❑ Two observations



Regressions between τ_{Ku} and τ_X , ϖ_{Ku} and ϖ_X : based on LUT

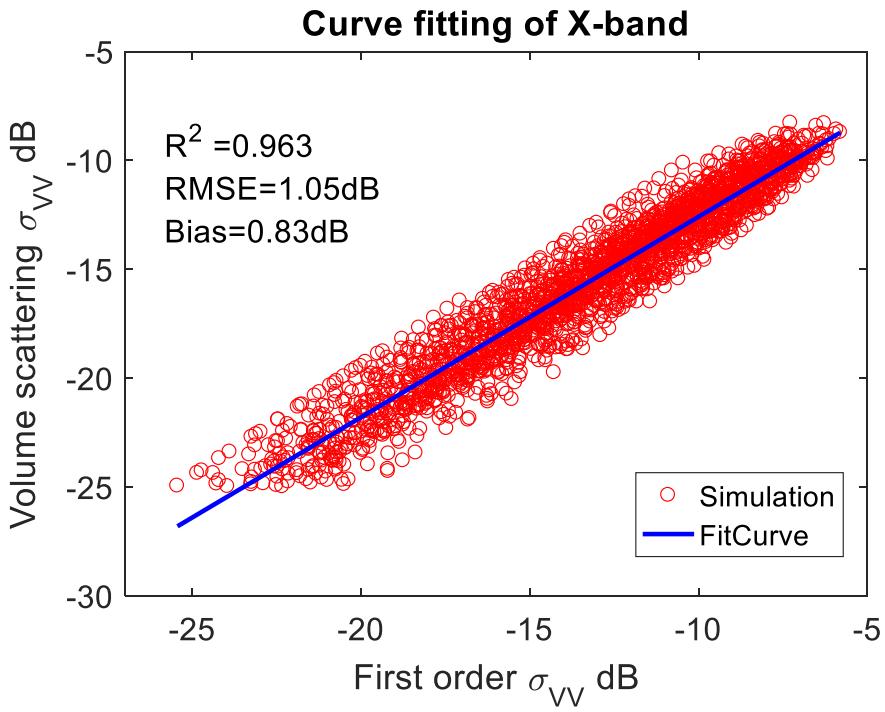


Correlation between (τ_X, τ_{Ku})

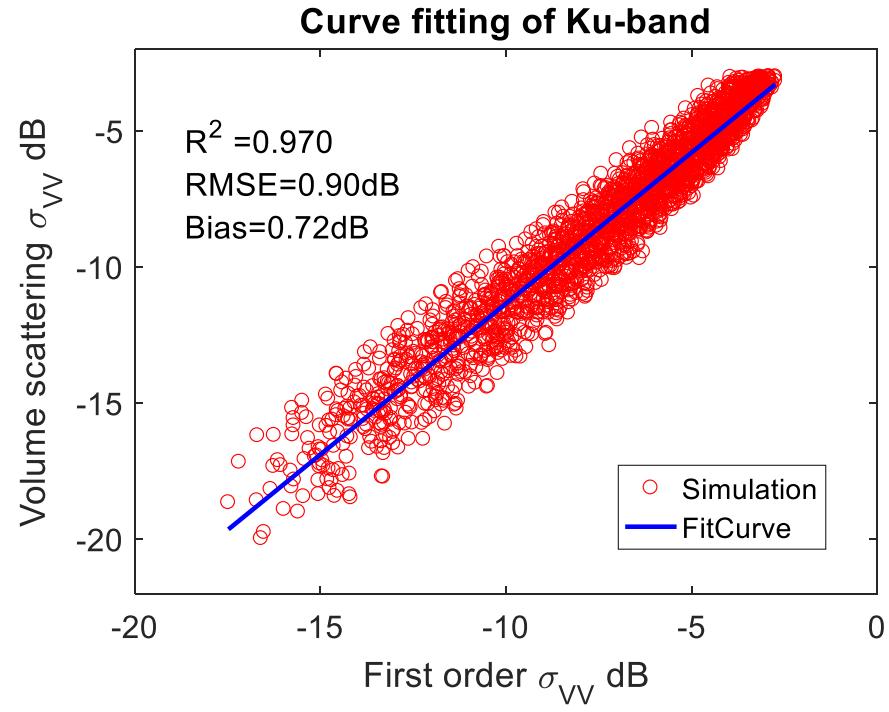


Correlation between (ϖ_X, ϖ_{Ku})

Regression between single and multiple scattering

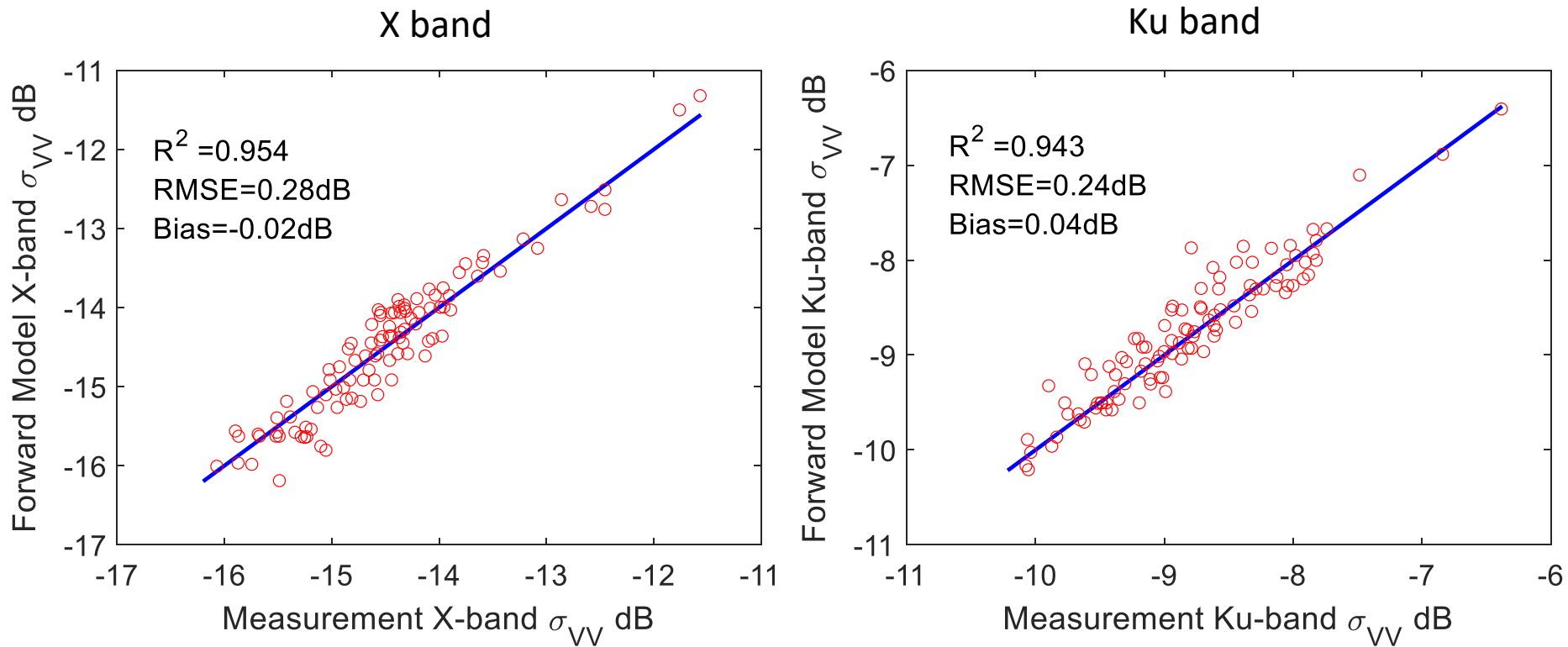


Backscatter for X band $\sigma_X \left(\sigma_X^{(1)}(\varpi_X, \tau_X) \right)$



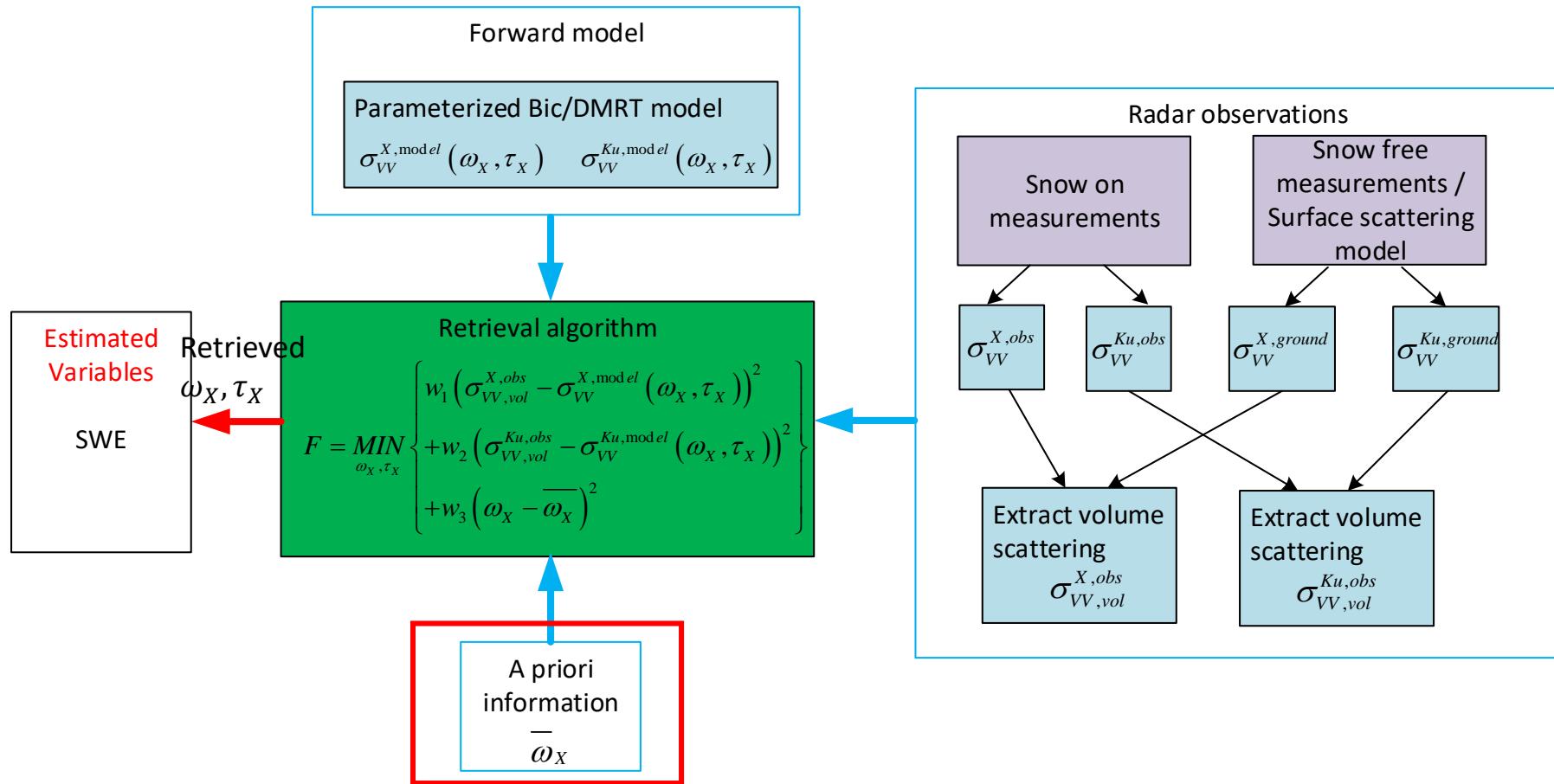
Backscatter for Ku band $\sigma_{Ku} \left(\sigma_{Ku}^{(1)}(\varpi_{Ku}, \tau_{Ku}) \right)$

Validation of parameterized Bic/DMRT model: Canada SnowSAR

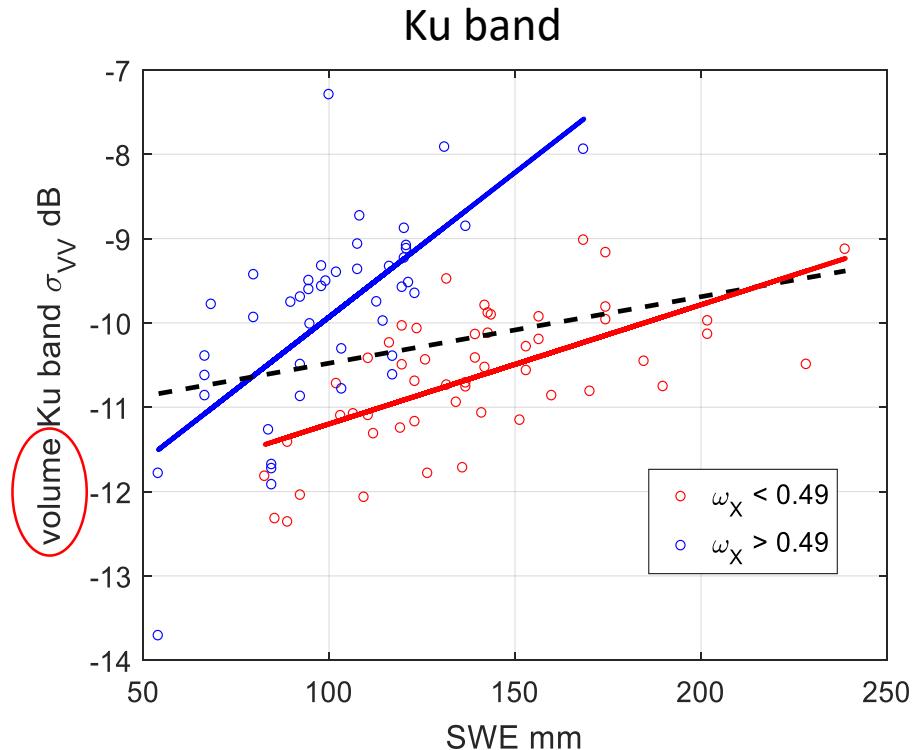
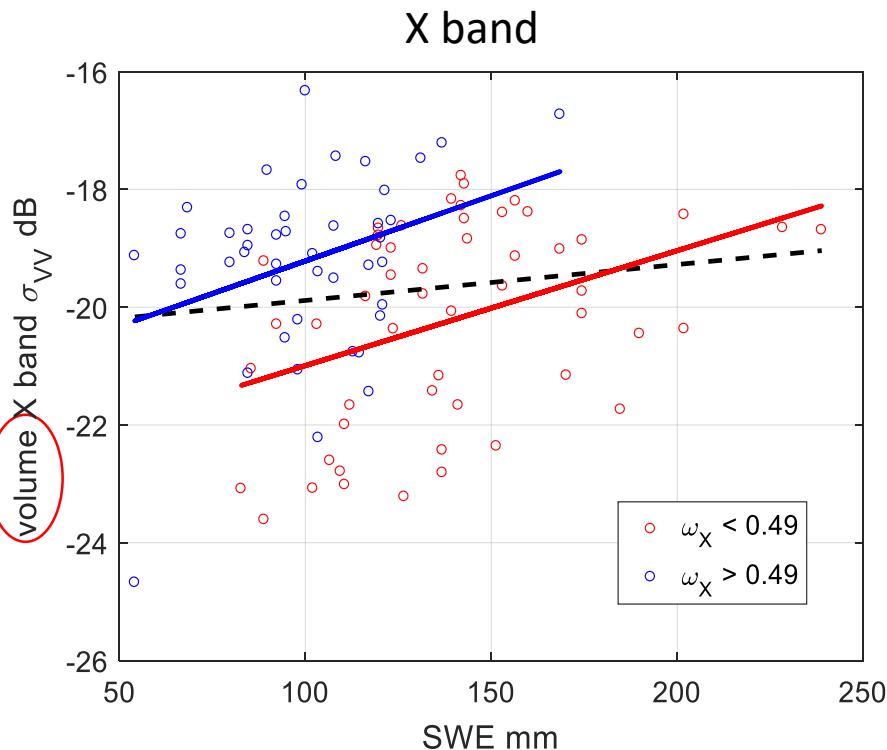


□ Good agreement: achieve RSME < 0.28dB

SWE retrieval algorithm flow chart



Classification: two classes of backscatter, Canada SnowSAR

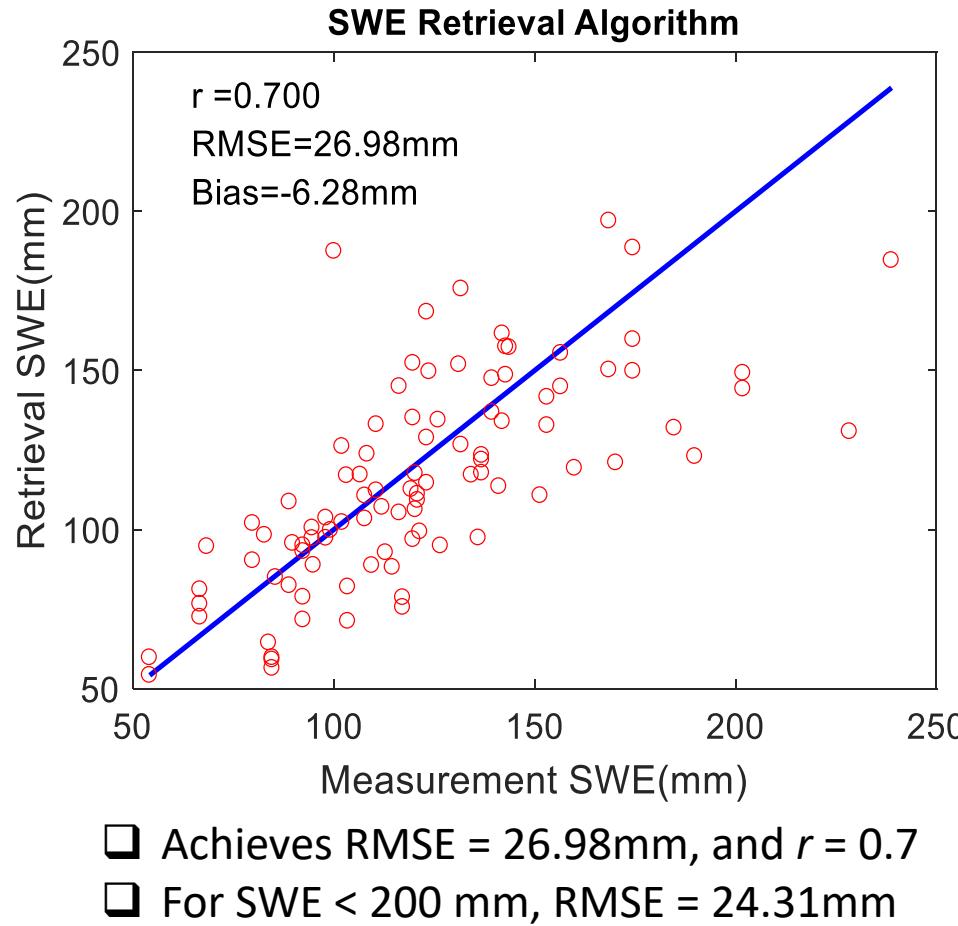


- Background scattering subtraction & backscatter classification w.r.t. ω_X**
enhances sensitivity of backscatter to SWE
- SWE doubles, Backscatter increases about 2-3dB

Radar datasets used

Dataset	Location	Date	Frequency	Polarization
Finland SnowSAR1	Sodankylä, Finland	Mar. 17th, 2011	X and Ku band	VV&HV
Finland SnowSAR2	Sodankylä, Finland	December 19 th , 2011 to March 24th, 2012	X and Ku band	VV&HV
Canada SnowSAR	Trail Valley Creek (TVC), the Northwest Territories, Canada	winter 2012~2013	X and Ku band	VV&HV

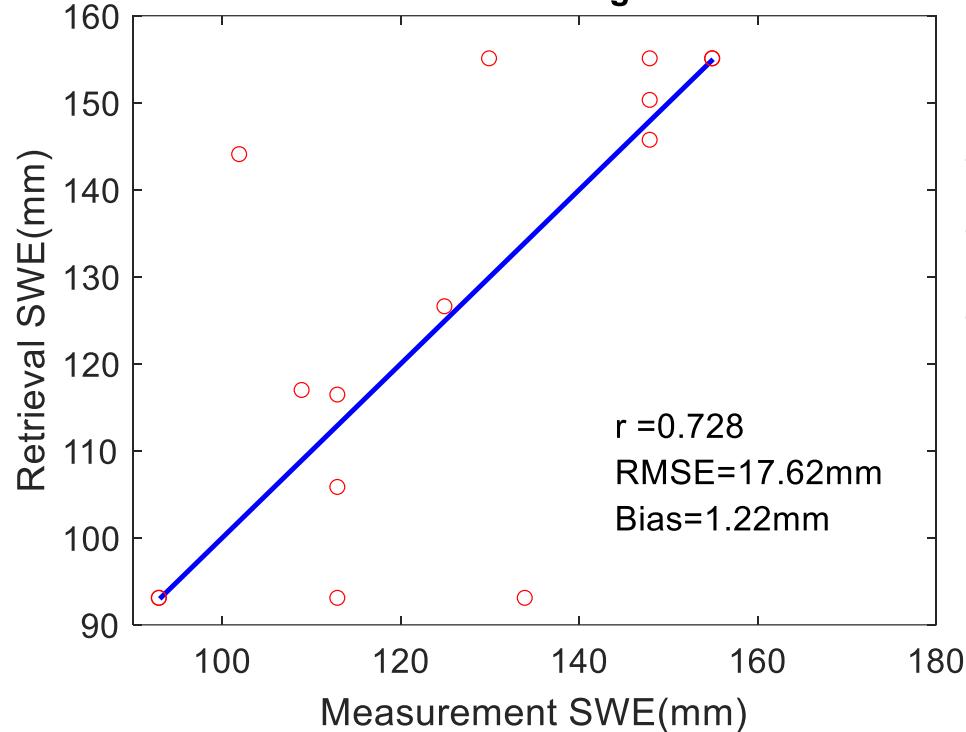
Performance of SWE retrieval algorithm: Canada SnowSAR



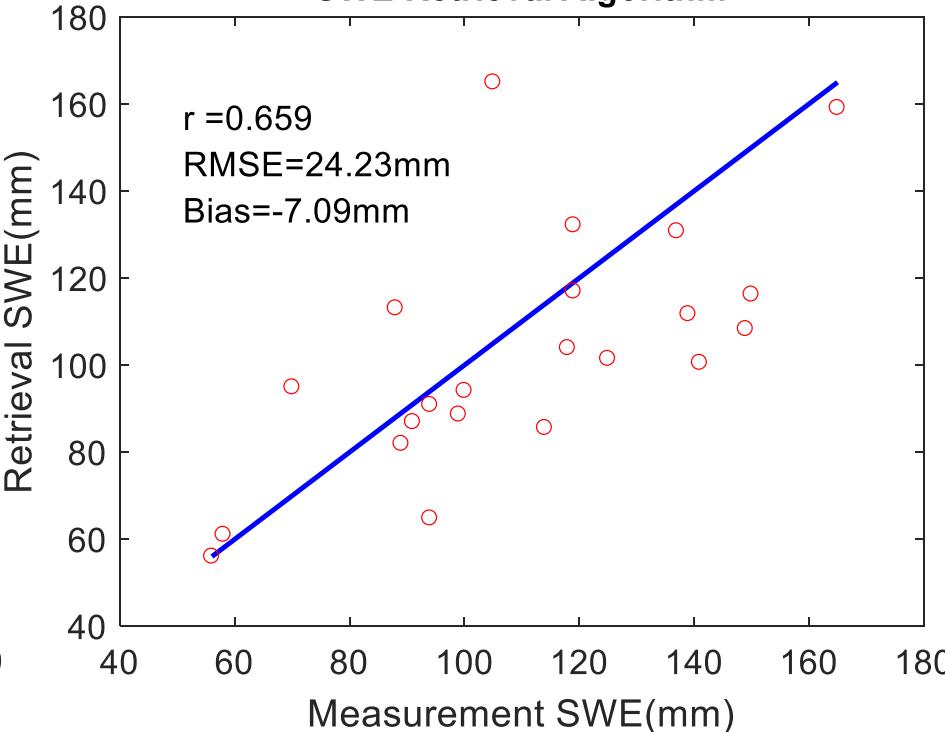
SCLP requirement: $\text{RMSE} < 20\text{mm}$ for $\text{SWE} < 200\text{mm}$
and $\text{RMSE} < 10\%$ of total SWE for $\text{SWE} > 200\text{mm}$

Performance of SWE retrieval algorithm: Finland SnowSAR1 and SnowSAR2

SnowSAR1
SWE Retrieval Algorithm



SnowSAR2
SWE Retrieval Algorithm



❑ Achieves RMSE = ~18 mm

❑ Achieves RMSE = ~24 mm

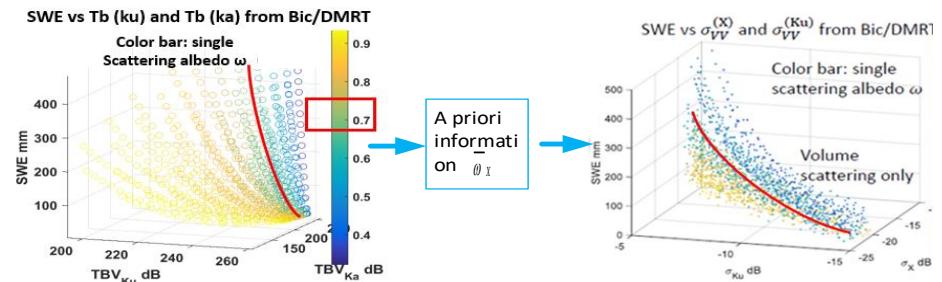
Methods to improve the algorithm

Better background
scattering subtraction

- Radar observation σ_{obs} from snow free conditions
- Polarimetry: volume / surface scattering decomposition
- Combine active and passive measurements to retrieve both soil and snowpack parameters

Better a priori estimate of ω_X
(or effective grain size)

- Solution 1: snow thermodynamics model with ancillary meteorological data
- Solution 2: combine active and passive microwave measurements



Summary

- A. Background scattering subtraction:
 - i. Affects more in X band than Ku band
 - ii. Volume backscatter sensitive to SWE
- B. Forward model: parameterized Bic/DMRT
 - i. Regression training: 2 observations vs. 2 unknowns (ω_X and τ_X)
 - ii. Validated against SnowSAR data
- C. Retrieval algorithm: $SWE \propto \tau_{a,X} = (1 - \omega_X)\tau_X$
 - i. A priori ω_X
 - ii. Classify backscatter w.r.t. ω_X restores its high sensitivity to SWE
 - iii. Performance: RMSE <30mm for SWE up to 300mm

Thanks for your attention!

Any question?